

Assignment - 2 k-NN for classification.

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```
#install.packages("readr")  
library(readr)  
#install.packages("lattice")  
library(lattice)  
#install.packages("caret")  
library(caret)
```

```
## Loading required package: ggplot2
```

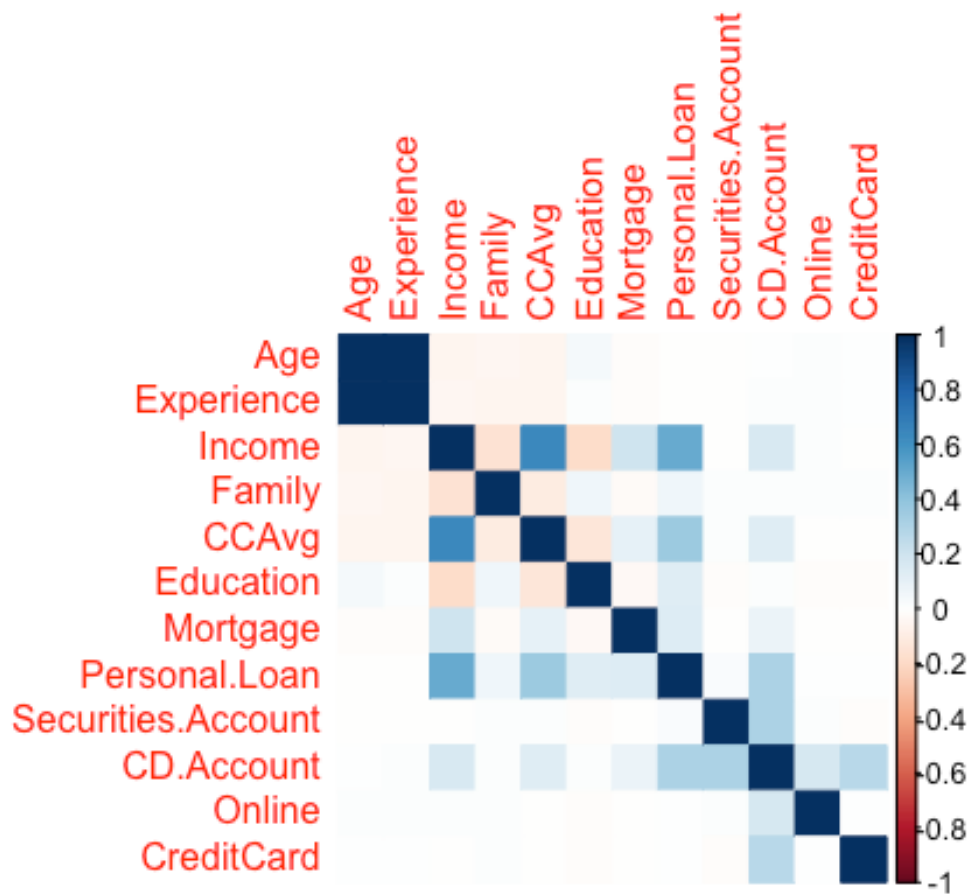
```
#install.packages("ISLR")  
library(ISLR)  
#install.packages("ggplot2")  
library(ggplot2)  
#install.packages("corrplot")  
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
#install.packages("fastDummies")  
library(fastDummies)  
#install.packages("FNN")  
library(FNN)  
#install.packages("plyr")  
library("plyr")  
#install.packages("gmodels")  
library(gmodels)  
#install.packages("ggplot2")  
library(ggplot2)
```

```
#Importing Data, Data visulization & Data Summary
```

```
options(stringsAsFactors = FALSE)  
UniversalBank <- read.csv("~/Desktop/FML/UniversalBank.csv")  
Universalbank_num <- UniversalBank[, c(2:4, 6:14)]  
corrplot(cor(Universalbank_num), method="color")
```



```
summary(Universalbank_num)
```

```
##      Age      Experience      Income      Family
##  Min.   :23.00   Min.   :-3.0   Min.    : 8.00   Min.    :1.000
## 1st Qu.:35.00   1st Qu.:10.0   1st Qu.: 39.00   1st Qu.:1.000
## Median :45.00   Median :20.0   Median : 64.00   Median :2.000
## Mean   :45.34   Mean   :20.1   Mean    : 73.77   Mean    :2.396
## 3rd Qu.:55.00   3rd Qu.:30.0   3rd Qu.: 98.00   3rd Qu.:3.000
## Max.   :67.00   Max.   :43.0   Max.    :224.00   Max.    :4.000
##      CCAvg      Education      Mortgage      Personal.Loan
##  Min.    : 0.000   Min.    :1.000   Min.    : 0.0   Min.    :0.000
## 1st Qu.: 0.700   1st Qu.:1.000   1st Qu.: 0.0   1st Qu.:0.000
## Median : 1.500   Median :2.000   Median : 0.0   Median :0.000
## Mean    : 1.938   Mean    :1.881   Mean    : 56.5   Mean    :0.096
## 3rd Qu.: 2.500   3rd Qu.:3.000   3rd Qu.:101.0   3rd Qu.:0.000
## Max.    :10.000   Max.    :3.000   Max.    :635.0   Max.    :1.000
## Securities.Account  CD.Account      Online      CreditCard
##  Min.    :0.0000   Min.    :0.0000   Min.    :0.0000   Min.    :0.000
## 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.000
## Median :0.0000   Median :0.0000   Median :1.0000   Median :0.000
## Mean    :0.1044   Mean    :0.0604   Mean    :0.5968   Mean    :0.294
## 3rd Qu.:0.0000   3rd Qu.:0.0000   3rd Qu.:1.0000   3rd Qu.:1.000
## Max.    :1.0000   Max.    :1.0000   Max.    :1.0000   Max.    :1.000
```

```
head(UniversalBank,10)
```

```
##      ID Age Experience Income ZIP.Code Family CCAvg Education Mortgage
## 1    1  25         1     49   91107      4   1.6         1         0
## 2    2  45        19     34   90089      3   1.5         1         0
## 3    3  39        15     11   94720      1   1.0         1         0
## 4    4  35         9    100   94112      1   2.7         2         0
## 5    5  35         8     45   91330      4   1.0         2         0
## 6    6  37        13     29   92121      4   0.4         2        155
## 7    7  53        27     72   91711      2   1.5         2         0
## 8    8  50        24     22   93943      1   0.3         3         0
## 9    9  35        10     81   90089      3   0.6         2        104
## 10  10 34         9    180   93023      1   8.9         3         0
##      Personal.Loan Securities.Account CD.Account Online CreditCard
## 1          0          1          0          0          0
## 2          0          1          0          0          0
## 3          0          0          0          0          0
## 4          0          0          0          0          0
## 5          0          0          0          0          1
## 6          0          0          0          1          0
## 7          0          0          0          1          0
## 8          0          0          0          0          1
## 9          0          0          0          1          0
## 10         1          0          0          0          0
```

```
#Convert Education to dummy variables
```

```
Universalbank_dummy <- dummy_cols(Universalbank_num, select_columns =
"Education")
```

```
#Splitting data Training : 60% , Validation : 40%
```

```
set.seed(1)
#splitting 60% of data into training & 40% of data into validation
Train_index <- createDataPartition(Universalbank_dummy$'Personal.Loan',
p=0.6, list=FALSE)
Training_data <- Universalbank_dummy[Train_index,]
Validation_data <- Universalbank_dummy [-Train_index,]
summary(Training_data)
```

```
##      Age      Experience      Income      Family
## Min.   :23.00  Min.   : -3.00  Min.    :  8.00  Min.    :1.000
## 1st Qu.:36.00  1st Qu.:10.00  1st Qu.: 39.00  1st Qu.:1.000
## Median :45.00  Median :20.00  Median : 63.00  Median :2.000
## Mean   :45.43  Mean   :20.19  Mean    : 73.08  Mean    :2.388
## 3rd Qu.:55.00  3rd Qu.:30.00  3rd Qu.: 98.00  3rd Qu.:3.000
## Max.    :67.00  Max.    :43.00  Max.    :224.00  Max.    :4.000
##      CCAvg      Education      Mortgage      Personal.Loan
## Min.    : 0.000  Min.    :1.00  Min.    :  0.00  Min.    :0.00000
## 1st Qu.: 0.700  1st Qu.:1.00  1st Qu.:  0.00  1st Qu.:0.00000
## Median : 1.500  Median :2.00  Median :  0.00  Median :0.00000
```

```
## Mean : 1.915 Mean :1.88 Mean : 57.34 Mean :0.09167
## 3rd Qu.: 2.500 3rd Qu.:3.00 3rd Qu.:102.00 3rd Qu.:0.00000
## Max. :10.000 Max. :3.00 Max. :635.00 Max. :1.00000
## Securities.Account CD.Account Online CreditCard
## Min. :0.0000 Min. :0.00000 Min. :0.0000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.00000 1st Qu.:0.0000 1st Qu.:0.0000
## Median :0.0000 Median :0.00000 Median :1.0000 Median :0.0000
## Mean :0.1003 Mean :0.05367 Mean :0.5847 Mean :0.2927
## 3rd Qu.:0.0000 3rd Qu.:0.00000 3rd Qu.:1.0000 3rd Qu.:1.0000
## Max. :1.0000 Max. :1.00000 Max. :1.0000 Max. :1.0000
## Education_1 Education_2 Education_3
## Min. :0.0000 Min. :0.000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.000 1st Qu.:0.0000
## Median :0.0000 Median :0.000 Median :0.0000
## Mean :0.4173 Mean :0.285 Mean :0.2977
## 3rd Qu.:1.0000 3rd Qu.:1.000 3rd Qu.:1.0000
## Max. :1.0000 Max. :1.000 Max. :1.0000
```

summary(Validation_data)

```
## Age Experience Income Family
## Min. :23.0 Min. :-3.00 Min. : 8.00 Min. :1.000
## 1st Qu.:35.0 1st Qu.:10.00 1st Qu.: 39.00 1st Qu.:1.000
## Median :45.0 Median :20.00 Median : 64.00 Median :2.000
## Mean :45.2 Mean :19.97 Mean : 74.81 Mean :2.409
## 3rd Qu.:55.0 3rd Qu.:30.00 3rd Qu.: 99.00 3rd Qu.:3.000
## Max. :67.0 Max. :43.00 Max. :218.00 Max. :4.000
## CCAvg Education Mortgage Personal.Loan
## Min. : 0.000 Min. :1.000 Min. : 0.00 Min. :0.0000
## 1st Qu.: 0.700 1st Qu.:1.000 1st Qu.: 0.00 1st Qu.:0.0000
## Median : 1.600 Median :2.000 Median : 0.00 Median :0.0000
## Mean : 1.973 Mean :1.882 Mean : 55.24 Mean :0.1025
## 3rd Qu.: 2.600 3rd Qu.:3.000 3rd Qu.: 97.25 3rd Qu.:0.0000
## Max. :10.000 Max. :3.000 Max. :617.00 Max. :1.0000
## Securities.Account CD.Account Online CreditCard
## Min. :0.0000 Min. :0.0000 Min. :0.000 Min. :0.000
## 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.000 1st Qu.:0.000
## Median :0.0000 Median :0.0000 Median :1.000 Median :0.000
## Mean :0.1105 Mean :0.0705 Mean :0.615 Mean :0.296
## 3rd Qu.:0.0000 3rd Qu.:0.0000 3rd Qu.:1.000 3rd Qu.:1.000
## Max. :1.0000 Max. :1.0000 Max. :1.000 Max. :1.000
## Education_1 Education_2 Education_3
## Min. :0.000 Min. :0.000 Min. :0.000
## 1st Qu.:0.000 1st Qu.:0.000 1st Qu.:0.000
## Median :0.000 Median :0.000 Median :0.000
## Mean :0.422 Mean :0.274 Mean :0.304
## 3rd Qu.:1.000 3rd Qu.:1.000 3rd Qu.:1.000
## Max. :1.000 Max. :1.000 Max. :1.000
```

```
#checking Frequency of personal Loan splited properly or not
count(Training_data$`Personal.Loan`)
```

```
##    x freq
## 1 0 2725
## 2 1   275
```

```
count(Validation_data$`Personal.Loan`)
```

```
##    x freq
## 1 0 1795
## 2 1   205
```

#Data Normalization

```
train.normalized.df <- Training_data
valid.normalized.df <- Validation_data
norm.values <- preProcess(Training_data[, 1:7], method=c("center", "scale"))
#Replacing columns with normalized values
train.normalized.df[, 1:7] <- predict(norm.values, Training_data[, 1:7])
valid.normalized.df[, 1:7] <- predict(norm.values, Validation_data[, 1:7])
```

#KNN Modeling

```
cl= as.data.frame(train.normalized.df[,8])
tnf = as.data.frame(train.normalized.df)
vnf = as.data.frame(valid.normalized.df)
dim(cl)

## [1] 3000    1

dim(train.normalized.df[,1:7])

## [1] 3000    7

dim(valid.normalized.df[,1:7])

## [1] 2000    7

knn_predict <- knn(tnf, vnf, cl=train.normalized.df$`Personal.Loan`, k =1)
head(knn_predict)

## [1] 0 0 0 0 1 0
## Levels: 0 1

knn_predict <- as.data.frame(knn_predict)
```

#assess Data to model

```
customer_df <- data.frame ("Age" =40, "Experience"=10, "Income"=84,
"Family"=2, "CCAvg"=2, "Education_1"=0, "Education_2"=1, "Education_3"=0,
"Mortgage"=0, "Securities Account"=0, "CD Account"=0, "Online" =1, "Credit
Card"=1)
```

```

dim(tnf)

## [1] 3000 15

dim(customer_df)

## [1] 1 13

customerClass <- knn ((tnf[, c(-6, -8)]), (customer_df), cl =
train.normalized.df$`Personal.Loan`, k = 1, prob = 0.5)

summary(customerClass) #CUSTOMER class is 1. Customer is likely to accept a
personal loan according to this model.

## 1
## 1

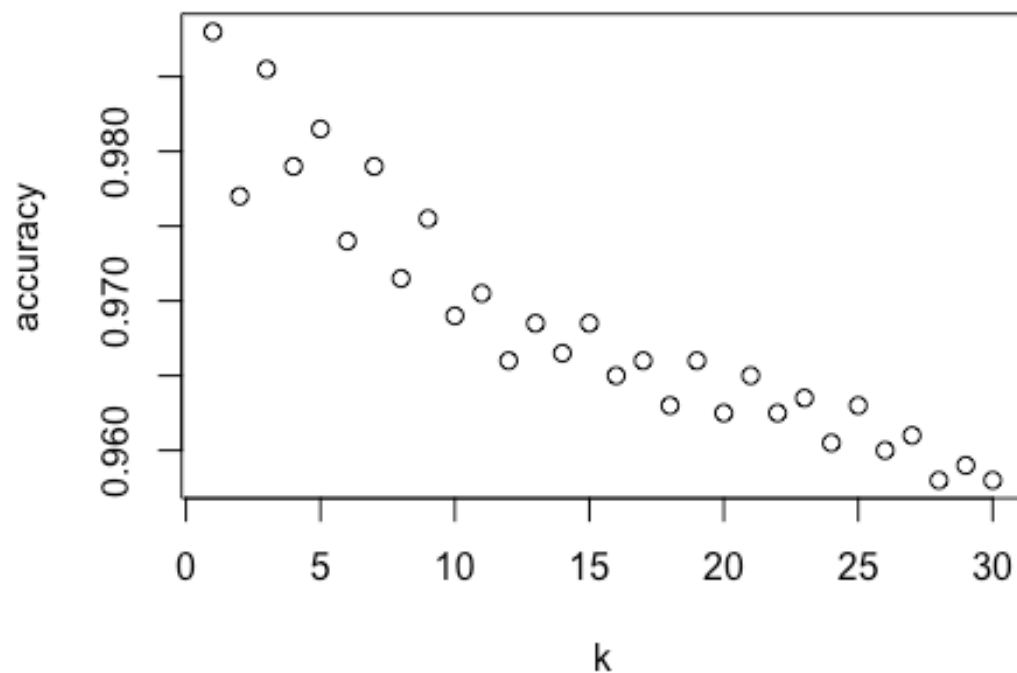
#Library(Lattice)
#Library(ggplot2)
#Library(caret)
accuracy.df <- data.frame(k= seq (1, 30, 1), accuracy = rep(0, 30))
for( i in 1:30) {
  prediction <- knn ( tnf, vnf, cl = train.normalized.df$`Personal.Loan`,
k = i)
  accuracy.df[i, 2] <- confusionMatrix ( as.factor (prediction), as.factor(
valid.normalized.df$`Personal.Loan`))$overall[1]
}
accuracy.df

##      k accuracy
## 1 1 0.9880
## 2 2 0.9770
## 3 3 0.9855
## 4 4 0.9790
## 5 5 0.9815
## 6 6 0.9740
## 7 7 0.9790
## 8 8 0.9715
## 9 9 0.9755
## 10 10 0.9690
## 11 11 0.9705
## 12 12 0.9660
## 13 13 0.9685
## 14 14 0.9665
## 15 15 0.9685
## 16 16 0.9650
## 17 17 0.9660
## 18 18 0.9630
## 19 19 0.9660
## 20 20 0.9625

```

```
## 21 21 0.9650
## 22 22 0.9625
## 23 23 0.9635
## 24 24 0.9605
## 25 25 0.9630
## 26 26 0.9600
## 27 27 0.9610
## 28 28 0.9580
## 29 29 0.9590
## 30 30 0.9580
```

```
plot(accuracy.df)
```



```
#Confusion Matrix
```

```
#library(gmodels)
```

```
valid_labels <-as.data.frame( vnf[,8])
```

```
#Model accuracy = TP+TN/Total= 99%, specificity= 99.7%, percision= 98%
```

```
CrossTable( valid_labels$`vnf[, 8]`, knn_predict$knn_predict, prop.chisq = FALSE)
```

```
##
```

```
##
```

```
## Cell Contents
```

```
## |-----|
## |                N |
## |      N / Row Total |
## |      N / Col Total |
## |      N / Table Total |
## |-----|
##
##
## Total Observations in Table:  2000
##
##
##      knn_predict$knn_predict
## valid_labels$`vnf[, 8]`      0      1 Row Total
## -----|-----|-----|
##              0      1795      0      1795
##              1.000      0.000      0.897
##              0.987      0.000
##              0.897      0.000
## -----|-----|-----|
##              1       24      181      205
##              0.117      0.883      0.102
##              0.013      1.000
##              0.012      0.090
## -----|-----|-----|
##      Column Total      1819      181      2000
##              0.909      0.090
## -----|-----|-----|
##
##
```

#Data Plinting into Training as 50% , Validation as 30% , Testing as 20%

```
set.seed(12)
Train_index2 <- createDataPartition(Universalbank_dummy$`Personal.Loan`,
p=0.50, list=FALSE)
Training_data2 <- Universalbank_dummy[Train_index2,]

CombinedValidation_test <- Universalbank_dummy [-Train_index2,]

Valid_index2 <- createDataPartition (CombinedValidation_test$`Personal.Loan`,
p=0.30, list=FALSE)
Validation_data2 <- CombinedValidation_test[Valid_index2,]
Test_data2 <- CombinedValidation_test[-Valid_index2,]
```

#Data Normalization

```
train.normalized.df2 <- Training_data2
valid.normalized.df2 <- Validation_data2
Test.normalized.df2 <- Test_data2
Combined_normalized2<-CombinedValidation_test
```



```

norm.values2 <- preProcess(Training_data2[, 1:7], method=c("center",
"scale"))

train.normalized.df2 [, 1:7] <- predict(norm.values2, Training_data2[,1:7])
# Replace columns with normalized values
valid.normalized.df2 [, 1:7] <- predict(norm.values2,
Validation_data2[,1:7])

Test.normalized.df2 [, 1:7] <- predict(norm.values2, Test_data2[, 1:7])

Combined_normalized2[, 1:7] <- predict(norm.values2,
CombinedValidation_test[,1:7])

```

#Modeling k-NN with validation data

```

#Library(FNN)
cl2= as.data.frame(train.normalized.df2[,8])
tnf2 = as.data.frame(train.normalized.df2)
vnf2= as.data.frame(valid.normalized.df2)
dim(cl2)

## [1] 2500    1

dim(train.normalized.df2[,1:7])

## [1] 2500    7

dim(valid.normalized.df2[,1:7])

## [1] 750    7

knn_predict2 <- knn(tnf2, vnf2, cl=train.normalized.df2$`Personal.Loan`, k
=1)
head(knn_predict2)

## [1] 0 0 0 0 0 1
## Levels: 0 1

knn_predict2 <- as.data.frame(knn_predict2)

```

#predicting KNN using validation and test data

```

cl2= as.data.frame(train.normalized.df2[,8])
tnf2 = as.data.frame(train.normalized.df2)
cnf3= as.data.frame(Combined_normalized2)
dim(cl2)

## [1] 2500    1

dim(train.normalized.df2[,1:7])

## [1] 2500    7

```

```

dim(Combined_normalized2[,1:7])

## [1] 2500    7

knn_predict3 <- knn(tnf2, cnf3, cl=train.normalized.df2$`Personal.Loan`, k
=1)
head(knn_predict3)

## [1] 0 0 0 0 0 0
## Levels: 0 1

knn_predict3 <- as.data.frame(knn_predict3)

summary(knn_predict3)

## knn_predict3
## 0:2295
## 1: 205

```

#Customer class

```

customer_df2 <- data.frame ("Age" =40, "Experience"=10, "Income"=84,
"Family"=2, "CCAvg"=2, "Education_1"=0, "Education_2"=1, "Education_3"=0,
"Mortgage"=0, "Securities Account"=0, "CD Account"=0, "Online" =1, "Credit
Card"=1)

dim(tnf2)

## [1] 2500    15

dim(customer_df2)

## [1] 1 13

customerClass2 <- knn ((tnf2[, c(-6, -8)]), (customer_df2), cl =
Combined_normalized2$`Personal.Loan`, k = 1, prob = 0.5)
#CUSTOMER class is 0. Customer is NOT likely to accept a personal loan
according to this model
summary(customerClass)

## 1
## 1

# k= 8 gives the highest accuracy percentage of 91%
accuracy.df2 <- data.frame(k= seq (1, 20, 1), accuracy = rep(0, 20))

for( y in 1:20){
  prediction2 <- knn (tnf2, cnf3, cl= Combined_normalized2$`Personal.Loan`,
k = y)
  accuracy.df2[y, 2] <- confusionMatrix ( as.factor(prediction2) ,

```

```
as.factor(Combined_normalized2$`Personal.Loan`))$overall[1]
}

## Warning in confusionMatrix.default(as.factor(prediction2),
## as.factor(Combined_normalized2$Personal.Loan)): Levels are not in the same
order
## for reference and data. Refactoring data to match.

## Warning in confusionMatrix.default(as.factor(prediction2),
## as.factor(Combined_normalized2$Personal.Loan)): Levels are not in the same
order
## for reference and data. Refactoring data to match.

## Warning in confusionMatrix.default(as.factor(prediction2),
## as.factor(Combined_normalized2$Personal.Loan)): Levels are not in the same
order
## for reference and data. Refactoring data to match.

## Warning in confusionMatrix.default(as.factor(prediction2),
## as.factor(Combined_normalized2$Personal.Loan)): Levels are not in the same
order
## for reference and data. Refactoring data to match.

## Warning in confusionMatrix.default(as.factor(prediction2),
## as.factor(Combined_normalized2$Personal.Loan)): Levels are not in the same
order
## for reference and data. Refactoring data to match.
```

```
## as.factor(Combined_normalized2$Personal.Loan): Levels are not in the same
order
## for reference and data. Refactoring data to match.

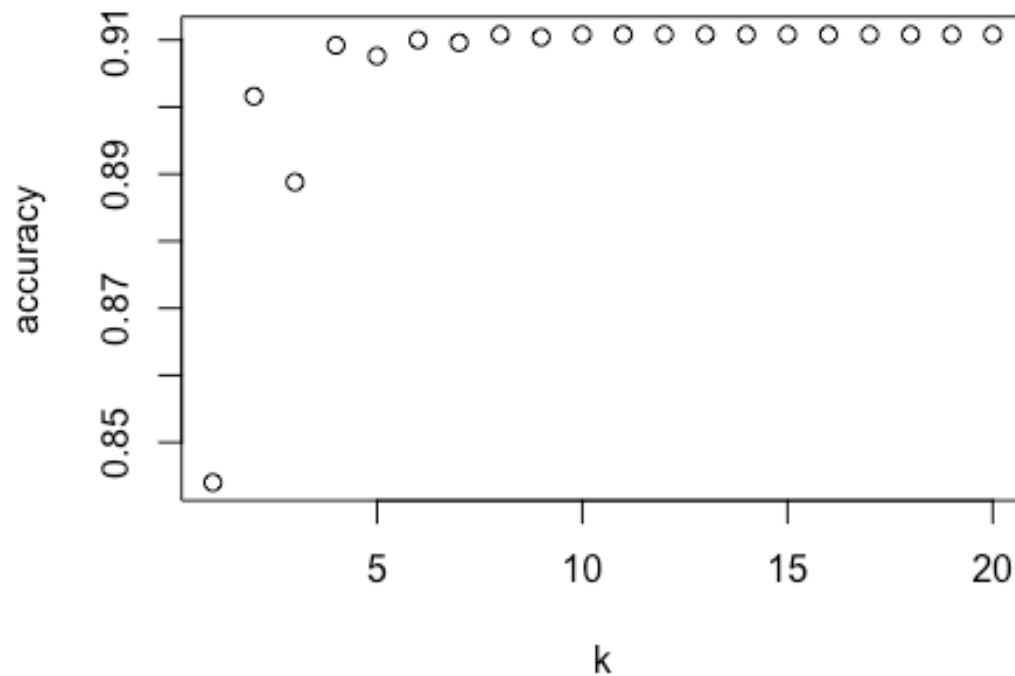
## Warning in confusionMatrix.default(as.factor(prediction2),
## as.factor(Combined_normalized2$Personal.Loan)): Levels are not in the same
order
## for reference and data. Refactoring data to match.

## Warning in confusionMatrix.default(as.factor(prediction2),
## as.factor(Combined_normalized2$Personal.Loan)): Levels are not in the same
order
## for reference and data. Refactoring data to match.

accuracy.df2

##      k accuracy
## 1  1  0.8440
## 2  2  0.9016
## 3  3  0.8888
## 4  4  0.9092
## 5  5  0.9076
## 6  6  0.9100
## 7  7  0.9096
## 8  8  0.9108
## 9  9  0.9104
## 10 10 0.9108
## 11 11 0.9108
## 12 12 0.9108
## 13 13 0.9108
## 14 14 0.9108
## 15 15 0.9108
## 16 16 0.9108
## 17 17 0.9108
## 18 18 0.9108
## 19 19 0.9108
## 20 20 0.9108

plot(accuracy.df2)
```



#Using only validation dataset

```
valid_labels2 <-as.data.frame( vnf2[,8])
```

```
CrossTable( valid_labels2$`vnf2[, 8]`, knn_predict2$knn_predict2,
prop.chisq = FALSE)      #Model accuracy = TP+TN/Total= 99%, specificity= 99.9%,
percision= 99%, sesitivity =93%
```

```
##
```

```
##
```

```
##      Cell Contents
```

```
## |-----|
## |                N |
## |      N / Row Total |
## |      N / Col Total |
## |      N / Table Total |
## |-----|
```

```
##
```

```
##
```

```
## Total Observations in Table:  750
```

```
##
```

```
##
```

```
## | knn_predict2$knn_predict2
```

```
## valid_labels2$`vnf2[, 8]`
## -----
##           0           1       Row Total
## -----
##           0           670           1           671
##           0.999           0.001           0.895
##           0.994           0.013
##           0.893           0.001
## -----
##           1           4           75           79
##           0.051           0.949           0.105
##           0.006           0.987
##           0.005           0.100
## -----
##           Column Total           674           76           750
##           0.899           0.101
## -----
##
##
```

#Using combined validation and test datasets

```
valid_labels2 <- as.data.frame(cnf3[, 8])
CrossTable( valid_labels2$`cnf3[, 8]`, knn_predict3$knn_predict3,
prop.chisq = FALSE )      #Model accuracy = TP+TN/Total= 99.9%, specificity=
99.9%, percision= 98.7%, sesitivity =91% This model give highest results.
```

```
##
##
##   Cell Contents
## |-----|
## |           N
## |   N / Row Total
## |   N / Col Total
## |   N / Table Total
## |-----|
##
##
## Total Observations in Table:  2500
##
##
##           knn_predict3$knn_predict3
## valid_labels2$`cnf3[, 8]`           0           1       Row Total
## -----
##           0           2274           3           2277
##           0.999           0.001           0.911
##           0.991           0.015
##           0.910           0.001
## -----
##           1           21           202           223
##           0.094           0.906           0.089
##           0.009           0.985
```

| | | | | |
|----|--------------|-------|-------|-------|
| ## | | 0.008 | 0.081 | |
| ## | ----- | ----- | ----- | ----- |
| ## | Column Total | 2295 | 205 | 2500 |
| ## | | 0.918 | 0.082 | |
| ## | ----- | ----- | ----- | ----- |
| ## | | | | |
| ## | | | | |